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A method for encoding image pixels, a method for processing images and a method for processing images
5 aimed at qualitative recognition of the object reproduced by one or more image pixels

The invention first relates to a method for encoding pixels of digital or digitized images, aimed
10 at making the information content of each pixel available to automatic image processing systems, particularly designed for image recognition with reference to the objects reproduced therein.

The term digital is intended to define an image
15 obtained by imaging apparatuses whose output is an image in digital format, i.e. digital cameras, Nuclear Magnetic Resonance imaging apparatuses, ultrasound imaging apparatuses and other imaging apparatuses.

The term digitized images is related to images
20 obtained by substantially analog systems, providing an analog image which is further scanned by means of devices known as scanners, regardless of whether the latter are hardware devices, i.e. devices for reading an image which is physically printed on a medium, or
25 software devices, i.e. designed to sample an image, provided in the form of set of signals and to turn it into digital signals.

The increasingly extensive use of digital images allows to have said images in the form of numerical
30 data, ready for further processing.

Regardless of whether the image is directly obtained digitally or acquired in an analog manner, by traditional systems, to be further digitized, e.g. scanned by digitizing devices, i.e. scanners, a digital
5 image is composed of a set of image dots, the so-called pixels, which may have different brightness conditions, i.e. different gray scale tones and, in color images, different colors. Each pixel of an image also has a well-defined position whereby the digital image may be
10 represented by a two- or three-dimensional matrix of elements $P_{i,j}$, each corresponding to a predetermined pixel of the pixel set that forms the image, the element $P_{i,j}$ being a variable which assumes the brightness and/or color value associated to the
15 specific pixel.

When the range is restricted to black and white images, the different pixel-associated brightness values are represented by a gray scale extending from black to white through several different intermediate
20 gray levels, whose number may be user-defined, based on the capabilities of the digitized imaging apparatus and/or of the display device.

In the case of a three-dimensional matrix, the discrete image unit element is generally referred to as
25 a voxel and the three-dimensional matrix is composed of elements $V_{i,j,k}$.

Therefore, from the technical point of view, a digital image has a unique equivalent in the form of a data matrix which forms a virtual image and, as such,
30 has a structure that is potentially adapted for image

processing by systems or methods which use algorithms, whether provided by software loaded in computers or by dedicated hardware for accomplishing specific functions on the image matrix.

5 Nevertheless, from a logical point of view, the information contained in each isolated pixel $P_{i,j}$ or voxel $V_{i,j,k}$ provides nothing but the simple indication of its brightness value, i.e. the gray scale value corresponding thereto, hence it has no meaning
10 wherefrom image information may be extracted, and only acts as data for controlling the display device, which may be and actually is handled during the imaging process to adjust the general aspect of the image, e.g. contrast and/or brightness and/or specific color
15 as defined based on user-selected functions, depending either on objective or subjective data.

 The image obtained thereby, as regards the shapes and characteristics of the objects reproduced by pixels in the different image portions derives from the
20 relation of each image pixel with the surrounding pixels. Therefore, in order to allow image processing to not only adjust the individual pixels to improve the quality of the displayed image, it is necessary to define the relations between each pixel and the pixels
25 around it. At present, no rule exists to determine such relations, except those defined on the basis of assumptions or presumptively defined rules, based on the specific characteristics of the objects reproduced by the image.

30 Therefore, the invention is based on the problem

of providing a method for encoding image pixels, which allows to account for the relations of each pixel with the pixels around it, substantially regardless of the peculiar characteristics of the object specifically reproduced in the image, i.e. a method that can be used to provide an image data set, particularly adapted for an image processing procedure aimed at recognizing at least some characteristics of the objects represented in the image, as well as of the shapes of these objects.

An additional object is to provide an encoding process as mentioned above, which is simple and requires neither complex processing steps, nor long processing times, and does not cause the hardware required to store the encoded data to be overloaded.

The invention achieves the above purposes by providing a method for encoding pixels of digital or digitized images, wherein each pixel of the pixel set which forms the image is uniquely identified with a vector whose components are given by the data of the pixel to be encoded and by the data of at least one or at least some or at least all of the pixels around the pixel to be encoded, which pixels are disposed within a predetermined subset of pixels included in the total set of pixels that form the image.

As a first approximation, the components of the pixel identifying vector are determined by selecting, as pixels surrounding the pixel to be identified, all the pixels that are directly adjacent to said pixel to be encoded.

The components of a pixel identifying vector may be also extended to at least one or at least some or all of the pixels which surround the pixels directly adjacent to the pixel to be encoded.

5 Theoretically, no limit is imposed to the number and position of pixels around the pixel to be identified, which may be used to define the components of the pixel identifying vector.

10 The components of the identification vector corresponding to the pixel to be identified and to the surrounding pixels are arranged in such a manner as to reflect the arrangement of the pixels within the pixel matrix which forms the image, with reference to a predetermined pixel reading sequence, for forming said
15 vector.

Particularly, the components of the identification vector, corresponding to the pixel to be identified and to the surrounding pixels are arranged in such a manner as to correspond to the distance relation of said
20 pixels with one another and with the pixel to be encoded, with reference to a predetermined pixel reading sequence, for forming said vector.

The components of the identification vector are arranged in such a manner that the pixel to be
25 identified has a central position which corresponds to the one taken in the image pixel set, obviously as related to the surrounding pixels, which pixel set has been selected for determining the identification vector components.

30 The methods includes the generation of an

identification vector for each pixel which forms the digital or digitized image.

By this arrangement, the virtual image composed of a data matrix which corresponds to a set of virtual
5 pixels, i.e. a set of data having the same position as that of real, actually displayed image pixels, is turned into a matrix in which each element has, at a certain pixel location, the identification vector therefor, which in turn has the numerical structure as
10 defined above.

Since the identification vector includes components given by the data associated to a certain predetermined number of pixels surrounding the pixel to be identified, the latter is defined not only by the
15 numerical value corresponding to the brightness thereof, but also by the numerical values that correspond to the brightness of the surrounding pixels, which were selected to form the identifying vector components. Hence, the matrix of pixels, i.e. of
20 brightness data associated to the pixels is changed into a set of vectors.

The pixel identifying vector may be also extended to other components, e.g. the values of the selected pixels and of the pixel to be identified at different
25 instants of time. This is advantageous when, for instance, different digital or digitized images of the same frame, acquired at different instants, are available. In this case, the vector will be associated to a succession of different sets of components,
30 comprising the value of the pixel to be identified and

the values of the selected pixels around it, each set being determined by an image acquired or referred to the same frame at different acquisition instants.

Preferably, the component sets are ordered within
5 the identification vector in a succession corresponding to the time sequence of capture thereof.

The above clearly shows that the encoding method of the invention, besides allowing to identify each pixel based on its numerical value and on the relation
10 of said pixel to a certain number of surrounding pixels, also extends this identification to the time variation of said pixel to be identified and to the time variations of the relations of said pixel to be identified to the selected surrounding pixels. Thanks
15 to the encoding method according to the invention, a numerical description may be provided for each image pixel, even for sequences of images representing moving objects, any change caused by the movement of the object being contained in the identification vector.

20 The method as described above may be easily implemented both for two- and three-dimensional imaging. In the latter case, the number of components of the identification vector considerably increases, in a cubic progression, if all the pixels which form the
25 increasingly distant shells around the pixel to be identified are to be accounted for.

The pixel selection pattern around the pixel to be identified and whose data shall form the components of the identifying vector may vary depending on actual
30 needs.

According to a further embodiment of the method for encoding image pixels, instead of considering as the relevant minimum image surface to be coded only one pixel a group of adjacent pixels may be considered. In
5 a square sub matrix of pixels for coding, the minimum image area to be coded can be formed by the four center pixels of the said sub matrix. In this case the at least one value assigned to the group of four pixels in the coding vector or matrix is formed by a linear or
10 non linear combination of the brightness of the said four adjacent pixels.

The value of the group of pixels to be coded as the minimum image area may be a combined value comprising several parameters each one considering for
15 example a linear or non linear combination of the brightness of the said four grouped pixels, such like a mean value of the brightness of the four grouped pixels, and statistical values such as brightness variance of the four grouped pixels. As numerical
20 values of the said grouped pixels considered a minimum image area also the colour values of the four grouped pixels might be considered if the image is a colour image and even the variation relatively to time of the said brightness and colour values. Also color values of
25 the grouped pixels might be in the form as a linear or non linear combination of the color values of the single pixels of the group and/or to values corresponding to statistical parameters such as variance or the like.

30 The variation of the brightness values and or of

the color values of the grouped pixels might be considered as the variations relatively to time of the linear or non linear combinations, such as mean values or other statistical parameter elatively to time.

5 In the identification or coding vector for such a minimum image area formed by a certain number, for example four grouped pixels, might comprise similarly to the above described identification vector the brightness and or color values and/or their variation
10 relating to time of a certain number of surrounding pixels.

 In this case of a minimum image area to be coded to alternatives might be choosen. In the first one each selected pixel surrounding the grouped pixels forming
15 the minimum image area to be coded is considered alone or per se.

 In a second alternative embodiment the selected surrounding pixels are grouped similarly to the ones to be coded thus forming surrounding minimum image zones
20 the values of this surrounding image zones being defined by linear or non linear combination or by other statistical functions of the brightness of the signle pixels being grouped as a minimum image surrounding zone.

25 The above disclosed embodiment has the advantage to reduce calculation time in the case the pixels are very small relatively to the subject reproduced by the image, so that within the area of only one pixel no variation of the structure of the imaged subject may
30 occur.

The invention also relates to an image processing method, particularly aimed at recognizing objects and/or object shapes, in an image in which pixels are encoded into identification vectors.

5 The method for processing digital or digitized images includes the following steps:

 Encoding different digital or digitized images into pixel identifying vectors, with image objects being identified for the selected types and qualities,
10 and associating each identification vector to the corresponding known type of object or to the known quality that are actually reproduced by each pixel identified by said identification vector, with reference to a predetermined list of different types or
15 qualities.

 Generating a teaching database for an expert processing system, comprising the binomials formed by said identification vectors and by the associated type or quality of the reproduced object;

20 Actually teaching the expert processing system, by entering the teaching database therein;

 Encoding the pixels of a not otherwise evaluated image by identification vectors associated to each pixel;

25 Entering said image pixel identifying vectors in the processing system to obtain, at the output of said processing system and as a result of the processing, the type or quality of the object represented by the image pixels.

30 The process of recognizing a quality or a type of

object and assigning it to a pixel identifying vector is not complex, because it only is a comparison. The difficulty lies in the great number of pixels, hence of the vectors to be compared, and in that the vectors
5 have a considerable number of components. When accounting for the processing time, however fast, of each step in which the identification vector for a pixel to be recognized is compared with the identification vectors for pixels included in the basic
10 knowledge or teaching database for the expert system, total times are extremely long and unacceptable in practice. This field of application is particularly suitable for particular processing algorithms and more precisely for the well-known neural networks.

15 Nevertheless, it shall be understood that, while the use of a neural network as a processing system is the preferred embodiment, the invention is not limited to said algorithm, but may address any type of algorithm for comparing the identification vectors of
20 the image pixels to be processed with the teaching database, such as a discretizing algorithm which makes a dull comparison and decides whether the identification vector belongs to one or the other type of object or feature amongst the various possibilities.

25 As will be apparent hereafter, the pixel encoding method according to this invention provides highly reliable and accurate results, i.e. higher than is currently expected.

These algorithms can rapidly converge toward the
30 correct solution or to a solution to be statistically

considered as the best possible solution, by also accounting for the considerable variances of the identification vectors for the same object type or quality.

5 The result provided by the expert processing system may be viewed by simply printing or displaying a list. As an alternative thereto or in combination therewith, the processing result may be highlighted by associating a color to each type or quality and by
10 representing the solution over the digital or digitized image with each pixel of the digitized image being assigned the color of the corresponding type or quality of the represented object, as determined by the expert processing system.

15 It shall be noted that the teaching step, based on either different images of the same frame at different times or on images of different frames or objects whose type or quality is one of the predetermined options, allows the expert processing system, particularly a so-
20 called neural network to learn what the aspect of the identification vector should be for a particular object or a particular quality with the highest variance of this aspect.

It shall be also understood that the recognition
25 of the reproduced objects and/or qualities is independent from the global processing of the image and that it is performed pixel after pixel with no reference to what the pixel set represents within the image.

30 These two characteristics are of basic importance.

By providing as much variance as possible for the identification vectors for pixels that represent a certain object type or quality, the processing system is allowed to recognize more accurately and reliably whether an identification vector, hence a pixel, belongs to a certain type of object or to a certain quality.

Pixel-based processing allows to substantially unlink the recognition of a pixel identifying vector as belonging to a certain object type or quality from the imaged subject.

The image processing method provides other advantages.

A first additional advantage consists in that the list of object types or qualities may be modified anytime, i.e. restricted or extended, without affecting the previous teaching process, by simple integration in the teaching database for the processing system. It is also possible to restrict image processing to only recognize some of the types or qualities of the imaged object among all the qualities or types of the teaching database, without affecting any further extension thereof.

Moreover, with use, the database including the knowledge acquired by the system may be increased, thereby improving knowledge, expertise, hence reliability and accuracy of the processing system.

Objective evaluation of pixel characteristics allows to avoid any human image recognition and interpretation errors.

By a targeted selection of the object types and/or qualities to be recognized and the generation of adequate teaching databases, the same processing system may be used to accomplish different functions.

5 Besides simply recognizing object types or qualities, it is possible to generate teaching databases which allow to correct defocused images or accurately overlay two digital images of the same subject, acquired with different imaging methods, e.g.
10 by Nuclear Magnetic Resonance imaging, ultrasound imaging or radiographic or photographic imaging.

Yet another advantage provided by the recognition method of this invention consists in allowing to limit image definition during acquisition, thereby obtaining
15 identical or even better results as regards the possibility to evaluate the acquired image thanks to a better and more accurate recognition allowed by the method, as compared with human eye potential. This provides an important advantage, a lower resolution
20 involving a reduced duration of imaging, e.g. by Nuclear Magnetic Resonance or by ultrasounds or other similar means. This not only allows to reduce the costs required for fast imaging and image reconstruction apparatuses, but also has positive implications, namely
25 for the comfort of the patient, who does not have to keep still for very long times.

A particular application of the image recognition method of the invention consists in the automatic recognition of tissue types from the diagnostic images
30 acquired by Nuclear Magnetic Resonance imaging,

ultrasound imaging, radiography, etc.

In this case, the method includes the following steps:

- generating a database for teaching the expert
5 processing system based on pixel identifying vectors
obtained by encoding well-known digital or digitized
diagnostic images, whose pixels have been previously
associated by an expert to a tissue type, which pixel
identifying vectors are uniquely associated to the
10 corresponding known tissue type;

- Teaching the expert system by entering the
teaching database therein, as well as the modes
associated to the processing algorithm type;

- Encoding an image or a time sequence of images
15 of the same frame, into pixel identifying vectors
according to the method of this invention, which image
or sequence is digitally acquired or has been digitized
but not interpreted;

- Entering pixel encoding vectors for the non
20 interpreted image entered into the processing system,
which processes the tissue type or the tissue quality
for each pixel identifying vector and displays the
result thereof.

Particularly, the result is displayed as color
25 assignments, where predetermined colors are assigned to
the different types of object or to the different
qualities of the pixels which have been found to belong
to an object type and/or quality.

It shall be noted that in this case the method has
30 no diagnostic function, but generates considerably

reliable indications for the physician or for the technical personnel responsible for the evaluation of the acquired image. No direct treatment suggestion is provided, but simply an indication of a type of tissue which is highly likely to be found in the image. The actual and total certainty of the result for final diagnosis requires both the image to be read and interpreted by qualified personnel and other cross-checks to be performed by other diagnostic methods.

Nevertheless, as will be more apparent from the following, it is not easy to read and interpret a diagnostic image, like a radiographic plate, an ultrasound image or a Nuclear Magnetic Resonance image, particularly when the diseases reproduced in the image have a very small extension. The instrument provided by this invention allows to reliably signal potential pathologic elements, while reducing the risk of misinterpretation, or preventing the same elements from being misinterpreted or even from not being seen by the physician or qualified personnel.

Further improvements of the invention will form the subject of the subclaims.

The characteristics of the invention and the advantages derived therefrom will appear more clearly from the following description of a non limiting embodiment, illustrated in the annexed drawings, in which:

Fig. 1 shows a simplified digital or digitized image pixel encoding scheme according to this invention, in a two-dimensional space.

Fig. 2 is a scheme as shown in Figure 1, referred to a three-dimensional imaging space.

Fig. 3 shows a scheme of the identification vector for the same pixel in a time sequence of digital or
5 digitized images.

Fig. 4 shows a block diagram of the inventive image processing method, which operates based on the pixel encoding method as shown in the previous figures.

Fig. 5 shows a data table and a chart of the image
10 processing results for the image as shown in Fig. 8 or in Fig. 11, obtained by Nuclear Magnetic Resonance imaging, and in which the processing system was operated with a teaching base only comprising two types of tissues, i.e. benign tumor tissues and malignant
15 tumor tissues.

Fig. 6 shows a data table, and a corresponding chart of the image processing results for a Nuclear Magnetic Resonance image, as seen in Fig. 8 or in Fig. 11, in which the processing system used a teaching base
20 comprising three types of tissues, i.e. benign tumor tissue, malignant tumor tissue and normal tissue.

Fig. 7 shows a data table and a corresponding chart of the image processing results for a Nuclear Magnetic Resonance image, as seen in Fig. 8 or in Fig.
25 9, in which the processing system operated with a teaching base comprising four types of tissues and an additional object type, i.e. benign tumor tissue, malignant tumor tissue, normal tissue, muscular tissue and background image.

30 Fig. 8 shows a first Nuclear Magnetic Resonance

image, in which the presence of a benign tumor tissue is outlined by a white ring.

Fig. 9 shows the same image as Figure 8, but processed with the method of the invention and with the mode in which only benign tumor tissues are recognized, in which the system has displayed the result on the originally acquired image by assigning a different aspect to the pixels which have been found to belong to the one or the other tumor tissue.

Fig. 10 shows the same image as Figures 8 and 9 once it has been processed according to this invention, the system having been taught to recognize the four types of tissues and the background, and the system having been prompted to display the originally acquired image by assigning different aspects to the pixels which have been found to belong to the different tissue types and/or to the background.

Figs. 11 and 12 are another Nuclear Magnetic Resonance image containing a malignant tumor tissue, as outlined by the white ring, and an enlarged image of said region containing the tumor tissue respectively.

Fig. 13 is the same image as Fig. 11, representing the result of image processing according to this invention, operated for recognition and differentiated display, on the same acquired image, of five different object types, including four tissue types (benign tumor, malignant tumor, normal and muscular) and a background type.

Fig. 14 shows the inventive processing method in combination with steps to display pixels corresponding

to a resolution below the human eye and to a resolution level with or worse than that of the human eye.

Figure 1 shows the inventive encoding method in a highly simplified manner, in the case of a two-dimensional digital or digitized image, i.e. consisting of a set of pixels (image unit elements).

The example shows a pixel of the pixel set, denoted as 5, which is designed to be encoded in such a manner as to make the information available for any type of treatment, particularly for processing.

In this case, the image may comprise any number of pixels and the steps of the method, as shown with reference to the pixel 5 of Figure 1 are executed for every pixel of the image. In the inventive encoding method, the pixels around the pixel are used to form an identification vector for the pixel 5. The surrounding pixels may be selected according to predetermined rules which may lead to different selections of surrounding pixels, as components of the identification vector, both as regards the number of the surrounding pixels to be selected as components of the identification vector and as regards the location of these surrounding pixels relative to the pixel to be encoded, in this case to the pixel 5.

One of the most obvious selections consists in using, as components of the identification vector for the pixel to be encoded, all the pixels directly adjacent to the pixel to be encoded, that is, in the notation referred to pixel 5 of Figure 1, the surrounding pixels 1, 2, 3, 4 and 6, 7, 8, 9.

In black and white or gray scale images, the value represented by each pixel is given by a brightness value of the corresponding pixel, i.e. a gray value in a gray scale extending from white to black through a certain number of intermediate levels, which may have a different number of gray tones, depending on the quality of the digital image with respect to the color resolution of the imaging apparatuses.

Depending on the type of color encoding, in color images, each pixel may also have one variable for indicating the color to be assigned thereto.

The example of the following figures will be limited to black and white or gray scale images, to better explain the steps of the method. The extension to pixel color indicating variables is an obvious step for those skilled in the art, which eventually involves the presence of a greater number of components in the identification vector.

Figure 1 shows the structure of the identification vector for a pixel with reference to pixel 5.

The vector comprises, in the same pixel indexing sequence within the pixel matrix, all the pixels which constitute the vector components, starting from pixel 1 and ending with pixel 9. In this case, the pixel 5 located at the center of the pixel matrix appears to occupy a central place in the sequence of the identification vector components.

The manner in which the pixels designed to form the identification vector for a pixel to be encoded are indexed is not relevant per se; what is important is

that the selection of this manner is consistently and accurately followed for all encoding processes, otherwise no comparison between two pixel vectors would be possible, since the vector components would have
5 different arrangement orders.

Therefore, it shall be noted that, for this encoding operation, the identification vector for pixel 5 does not only contain gray-scale brightness, i.e. pixel aspect information about the pixel to be
10 identified, but also brightness information about the pixels around it.

This vector structure is based on the acknowledgement that the content of an image is not recognizable based on the aspect of an individual
15 pixel, but based on the relation between the aspect thereof and the aspect of the surrounding pixels. In practice, each image dot is not important per se, unless it is evaluated with reference to the aspect of the surrounding dots or areas. Even from the visual
20 point of view, what is shown in an image is recognized on the basis of a relative evaluation between the different areas of the image.

As mentioned above, the selection of the surrounding pixels to create the identification vector
25 for the pixel to be encoded is not governed by any specific rule.

For example, it is possible to increase the number of surrounding pixels to be accounted for to generate the identification vectors, by using, as vector
30 components, at least some or all of the pixels of pixel

rings surrounding the central pixel to be encoded, at increasing distances from the central pixel to be encoded.

As in the illustrated example, it is possible to
5 account for all or some of the pixels of the pixel ring which externally surrounds the pixel ring directly adjacent to the central pixel 5 i.e. the pixels 1, 2, 3, 4, 6, 7, 8, 9.

In this case, the number of vector components
10 increases drastically, and overloads the identification vector processing conditions. If the identification vector for pixel 5 is arranged to comprise, for instance, all the pixels that externally surround the illustrated 3x3 pixel matrix, the number of
15 identification vector components increases from 9 to 25 components.

In this case, processing might obviously provide a more precise solution.

Furthermore, said additional pixels at a longer
20 distance from the pixel to be encoded may be suitably weighted, possibly also in a different manner from each other, to attenuate the effect thereof on the identification vector.

Fig. 2 shows the situation of a three-dimensional
25 image, in which the central pixel 14 is encoded by an identification vector which has, as components, the values of all the pixels directly surrounding it and subtending a 3x3x3 cube, whereby it includes 27 components.

30 The considerations proposed for the two-

dimensional embodiment also apply to the three-dimensional embodiment. This shows more clearly the overloaded processing of the identification vector when other pixels are accounted for, e.g. those of a three-dimensional cubic pixel shell which encloses the illustrated cubic set. In this case, the number of components increases in a cubic progression, from 27 pixels to 125 pixels, if all the pixels of a 5x5x5 cube are considered in the identification vector.

10 According to an additional characteristic of the invention, the pixel vector encoding method allows to also integrate the behavior through time of the pixel under examination in the identification vector, when a sequence of images of the same frame are available.

15 The sequence of images may be composed, for instance of frames of a motion picture or of individual images of the same frame as taken at successive instants. An example of imaging of the same frame at successive instants consists in diagnostic ultrasound imaging of contrast agent perfusion. In this case, the perfusion of contrast agents pushed by the flows of vascular circulation is imaged by injecting contrast agents in the anatomic part under examination at the instant T_c , and by subsequently imaging the same part at predetermined time intervals. Time variations of the image allow to check the presence of contrast agents, after a certain period from the injection instant. These images may provide useful deductions and/or information to check the presence of vascular and/or tumor diseases.

In the above case, the recognition of the reproduced object is not only based on the aspect thereof, but also on the time variation of said aspect. Therefore, the pixel vector encoding process aimed at
5 including, in the identification vector for each pixel, all the data characterizing the quality or type of the object reproduced by an image pixel must especially account for the time variation of the encoded pixel.

In this case, the identification vector for a
10 predetermined pixel, e.g. pixel 5, with reference to the embodiment of Figure 1, contains a set of 9 components, relating to pixels 1 to 9 in the proper time sequence, for each instant whereat the corresponding image has been captured.

15 The illustrated embodiment show six images of the same frame as acquired at the instants $T=0$, $T=1$, $T=2$, $T=3$, $T=4$, $T=5$. Moreover, the embodiment has been developed with reference to the acquisition of a sequence of ultrasound images of the same anatomic part, performed
20 after the injection of contrast agents. The instant TC whereat contrast agents are injected is denoted by the arrow TC.

The identification vector accounts for pixels 1 to 9 of the three images acquired at the instants $T=0$,
25 $T=1$, $T=2$ and in these conditions the identification vector already has 27 components. When all the 6 images of the time sequence are accounted for, the components will increase to 54. This only relates to an encoding process which accounts for a 9 pixel matrix, in which
30 the pixel to be encoded is the central pixel. When the

encoding process is intended to include the pixels surrounding the pixel to be encoded in a 5x5 pixel matrix, i.e. having 25 components of the identification vector for each instant, provided that all the images
5 of the sequence are considered, each identification vector will have 150 components.

In the case of a three-dimensional image as shown in the example of Fig. 2, in which the identification vector for a pixel, with surrounding pixels of a cubic
10 3x3x3 pixel space, has 27 components, the encoding vector according to the example of Figure 3 would include, for all six images of the image time sequence, 162 components. If encoding is extended to a 5x5x5 pixel three-dimensional space, the components of the
15 identification vector for the pixel will increase to 750.

It shall be noted that while the inventive encoding method may be followed in a simple and fast manner, it allows to identify the characteristics of an
20 image pixel through its value as well as in relation with its surrounding pixels, and also with reference to time variations of the pixel to be encoded and of the surrounding pixels.

It shall be further noted that this encoding
25 method only accounts for pixels, and of a substantially restricted examination field, which is independent from the subject of the image and of the encoding purpose.

Obviously, encoding times for an image or a sequence of images strictly depend on the image size,
30 in terms of number of pixels.

According to a further embodiment of the present pixel coding method, instead of considering an image formed by single pixels, for coding purposes a minimum image area of different size may be defined which is
5 formed by a more than one adjacent pixel.

So referring to the above described method and particularly to figure 1 to 3, with 5 in fig. 1 and 14 in fig. 2 there is not identified a single pixel but the said minimum image area formed by a certain number
10 of pixels, for example by a matrix of four adjacent pixels or of 9 adjacent pixels.

In this case the value of the said minimum image area formed by grouped adjacent pixels may be calculated as a linear or non linear combination and as
15 the value of one or more statistical functions of the of the brightness and or color values of the single pixels grouped together for forming the said minimum image area.

A simple example would consider the mean of the
20 brightness values of the single pixels forming the minimum image area the variance thereof. To this the information relating to the color of the pixels could be add also in the form of a mean and or of a variance value. Furthermore also in this case the time
25 dependency of the values relating to brightness and/or color could be add similarly as disclosed above relating to fig. 3. In this case the time dependency of the combination and of the statistical function of the brightness and or color values of the single pixels
30 will be considered, i.e. the variation relatively to

time of the mean value of the brightness or color values of the single pixels forming the minimum image area and or the variation relatively to time of the value of the statistical function.

5 As selected neighborhood pixels for the pixel vector identifying the said minimum image area, there might be the choice between two alternatives.

 In one alternative the selected pixels of the neighborhood are considered per se.

10 In the second alternative also the neighborhood pixels are grouped together for forming selected minimum image neighborhood areas. To explain more clearly the idea, if figure 1 or 2 are considered the image area indicated as pixel 1 to 4 and pixel 6 to 9
15 in the example of fig. 1 are not single pixels as described in the previous example but are formed by a certain number of adjacent pixels. This certain number of pixels might be the same as for defining the minimum image area to be coded or it might be a different
20 number of pixels as the one grouped for forming the minimum image area to be coded.

 The advantage of such a coding method considering minimum image area comprising more than one pixel will speed up the coding computational time and is useful
25 and will give not rise to errors particularly in the case the subject being imaged does not show any structural change within an area of several pixels, or the image resolution is so low that no structural variation can be sensed and shown within the area for
30 two or more adjacent pixels forming the image.

In the strictly diagnostic field, for instance, in which, due to practical and economic reasons, the number of image pixels is generally limited to 256 x 256 pixels for a two-dimensional image, encoding is
5 extremely fast. In this cases a coding according to the method as disclosed with reference to figures 1 to 3 may be used without suffering by a too high duration of the coding procedure.

The encoding method as disclosed above, which
10 includes the information about the aspect of a pixel with reference to the surrounding pixels and/or the time variations thereof, allows image processing methods which substantially require a recognition of the object type or of the quality of what is reproduced
15 by the pixel, and allow to automatically recognize the type of object and/or the quality thereof by the same processing system or software.

Figure 4 is a block diagram of a method of processing digital or digitized images, operating on
20 the basis of the previously described encoding method.

The processing method includes two steps: teaching the processing system and processing.

Processing is performed, per se, by an algorithm which is basically an algorithm for executing
25 comparisons between a database that includes a certain number of identification vectors for pixels associated to the type of object or to the quality corresponding thereto and the identification vectors for pixels of an image to be processed.

30 As a result of the comparison between the

identification vector for each pixel of the image to be processed and the identification vectors for the pixels included in the database, the comparison algorithm assigns to each pixel encoding vector within the image
5 to be processed, or to a predetermined portion thereof, the most appropriate or probable or the closest type of object or quality of the pixel identifying vector included in the database.

The processing algorithm may be a simple
10 discriminating algorithm, for instance an LDA algorithm (Linear Discriminant) (S.R. Searle, 1987, Linear Models for unbalanced data, New York, John Wiley & Sons) or a more complex algorithm, e.g. neural networks. The image processing procedure to be used is a typical
15 application for neural networks, i.e. an application in which a very great number of typically simple operations is required, and which finds no exact numerical solution due to the considerable number of identical processing steps to be performed. In
20 practice, a dull execution of the steps for comparing identification vectors for the image pixels to be processed, with the identification vectors for the pixels of the reference database, would require so lung computing times as to be unacceptable.

25 A number of neural networks might be used, for instance those known as: MetaGen1, MetaGen, MetanetAf, MetaBayes, MetanetBp, MetanetCm (M.Buscema(ed), 1998, SUM Special Issue on ANNs and Complex Social Systems, Volume 2, New York, Dekker, pp 439-461 and M.Buscema
30 and Semeion Group ,1999, Artificial Neural Networks

and Complex Social Systems [in italian], Volume 1, Rome
, Franco Angeli, pp 394-413 and M.Buscema, 2001, Shell
to program Feed Forward and Recurrent Neural Networks
and Artificial Organisms, Rome, Semeion Software n.12,
5 ver 5.0), TasmSABp, TasmSASn (M.Buscema and Semeion
Group ,1999, Artificial Neural Networks and Complex
Social Systems [in Italian], Volume 1, Rome , Franco
Angeli, pp. 440-464), FF-Bm, FF-Bp, FF-Cm, FF-Sn et al.
(D. E. Rumelhart, G. E. Hinton, and R. J.
10 Williams,1986, Learning representations by back-
propagating errors,Nature,23: 533-536; M.Buscema, 2000,
Squashing Theory and Contractive Map Network, Rome,
Semeion Technical Paper n. 23i -23e; M. Buscema, 1995,
Self-Reflexive Networks. Theory, Topology,
15 Applications, in Quality & Quantity, Kluwer Academic
Publishers, Dordrecht, The Netherlands, vol.29(4),
339-403, November). The publications describing the
above neural networks shall be considered as a part of
this invention.

20 The teaching step consists in generating a
database of pixel identification vectors which are
uniquely associated to the type of object or quality
reproduced by pixels of digital or digitized images
which are encoded as described above, and are
25 interpreted on the basis of visual operations performed
by qualified personnel. The identification vector of
each pixel is associated to the type of object or
quality of what is reproduced by the pixel, a list of
object types or qualities of interest having been
30 previously defined, consistent with the typical

subjects of the digital or digitized images used for teaching, hence for generating the knowledge database to be provided to the processing algorithm.

5 The knowledge database for teaching the processing algorithm is provided or allowed to be accessed by the processing algorithm, depending on the specific teaching mode of the selected processing algorithm.

10 At the end of the step for teaching the processing algorithm, a digital or digitized image of an image subject is encoded with the above described method, in a manner compatible with those used in the images designed to form the knowledge database and a list of object types or qualities is defined, among those included in the knowledge database of the processing
15 algorithm. The processing algorithm substantially compares the identification vectors of the individual pixels generated by the encoding process, and assigns the most probable object type or quality of the reproduced object, to each pixel.

20 The different indications of object types or qualities associated to each identification vector for image pixels are then displayed by printing lists and/or by differentially highlighting, e.g. by colors, the pixels of the image to be processed directly on the
25 image.

Depending on the total number of pixels and on the accuracy required by the list of object types or qualities reproduced in the individual pixels, either subsets of types or qualities or all the types and/or
30 qualities may be selected.

The digitized or digital images may be two-dimensional or three-dimensional with reference to what has been described for the encoding method, or may consist each of a sequence of images of the same frame,
5 as acquired at different instants.

In greater detail, Figure 4 shows the two teaching and processing steps in which 10 denotes a set of digital or digitized images, both individually and in the form of image sequences. 11 denotes the procedure
10 of encoding each pixel of said images into the corresponding identification vector. 12 denotes the step of uniquely associating the object quality or type as reproduced by each pixel to the corresponding identification vector based on the list of
15 predetermined object types or qualities 13 and 14 denotes the reference or teaching database for the image processing algorithm.

In the processing step, a digital or digitized image or a set of said images, such as a sequence of
20 images of the same frame, denoted as 18, is subjected to a step of pixel encoding into identification vectors, denoted as 19, and the identification vectors are provided to the processing algorithm 17, which is also supplied with a list of types or qualities
25 specifically sought for and included in the list 13 wherewith the teaching database 14 was prepared to be accessed by the processing algorithm 17. The processing algorithm assigns to each identification vector for the pixels of the image/s 18 an object type
30 or a quality and the identification vectors are decoded

in 20 into the corresponding pixel, the latter being assigned any pixel aspect changes uniquely related to the type associated thereto, for instance a color or the like. The pixels marked thereby are displayed on the screen, e.g. over the original image/s and/or a list of the identification vectors for the pixels of the digital image/s to processed is printed, and/or the image displayed on the screen is printed.

As an alternative thereto, the data provided from the algorithm may be used for further processing, based on the recognition of the object qualities or types reproduced by the individual pixels thanks to the processing algorithm.

Any further processing or handling of the data provided by the algorithm may be performed by the algorithm itself or by other types of algorithms, depending on the desired functions or handling purposes.

Hence, for example, the above processing method may be used to simply recognize objects, or qualities or conditions or statuses of the objects reproduced by pixels.

This type of processing is advantageously used in the medical field, as an automatic support to reading and interpretation of diagnostic images, particularly radiographic images, ultrasound images, Nuclear Magnetic Resonance images, or the like.

As an alternative thereto or in combination therewith, the method of the invention may be used to recognize shapes or types of objects in images with the

same subject and substantially the same frames, but being shot or acquired with different methods. In this case, each of the images of the same subject and showing substantially the same frame may be processed
5 with the processing method of the invention, whereupon the pixels of the different images, having substantially identical positions therein and being associated to the same object type or object quality are shown in overlaid positions, thereby providing an
10 image which contains the details of the same subject, as imaged with the three methods. This may be advantageous to integrate into a single image, details that may only be recognized and reproduced with some of the acquisition or imaging techniques or modes, as well
15 as details that may be only recognized and imaged with other acquisition or imaging techniques.

Similarly, the processing method may be used for image correction, e.g. to accurately correct defocused images. Here, by providing an adequate teaching
20 database including the identification vectors of defocused images with a uniquely associated object type or quality reproduced by the corresponding pixel, the inventive method may be used to generate a focused image, by identifying the pixels which reproduce
25 unfocused borders and removing or modifying them to obtain the focused image.

The two applications as described above may be obviously used in the medical field.

Particularly, by using the processing method to
30 obtain overlaid images substantially of the same frame

of the same subject, all data obtained by different imaging techniques, e.g. ultrasound, radiographic and MR imaging may be integrated into a single image.

It shall be noted that the possibility to
 5 recognize the different types or qualities of an object allows to integrate the information achieved by different imaging techniques even when the latter provides images of different frames of the same subject, while obviously accounting for the relative
 10 arrangement of said different frames in space.

Figures 5 to 13 show the results of an embodiment of the inventive method as applied to the medical field and to the purpose of supporting the diagnostic activity of the physician.

15 Particularly, the example as shown in Figs. 5 to 13 relates to the use of the processing method for selective recognition of different types of tissues in diagnostic Nuclear Magnetic Resonance images.

EXAMPLE 1 (Fig. 5)

SUBJECT	BREAST
IMAGING METHOD	NUCLEAR MAGNETIC RESONANCE
PURPOSE	RECOGNITION OF TISSUE TYPES
TISSUE TYPES	1. BENIGN TUMOR 2. MALIGNANT TUMOR
PIXEL ENCODING	3 X 3 PIXEL MATRIXES THE ENCODED PIXEL IS THE CENTRAL PIXEL
IMAGE DEFINITION	256x256 PIXEL

20

In the example 1, a teaching database for the image processing algorithm is generated to recognize two types of tissues, i.e. benign tumor and malignant tumor in the breast region.

5 A predetermined number of Nuclear Magnetic Resonance images of the breast region of patients who have been diagnosed a malignant breast tumor and of patients who have been diagnosed a benign breast tumor are pixel encoded according to the method described
10 above. The identification vectors for the pixels have, as components, all the surrounding pixels in a 3x3 pixel matrix, in which the pixel to be encoded is the central pixel (Fig. 1).

15 The identification vector for each pixel is assigned the type of tissue reproduced by the pixel in the image.

20 Therefore, the teaching database for the image processing algorithm contains identification vectors of image pixels relating to two tissue types, i.e. malignant tumor tissues of the breast region and benign tumor tissues of the breast region.

The following algorithms composed of so-called neural networks are used as processing algorithms.

PROCESSING ALGORITHM
MetaGen1
MetaGen
MetanetAf
FF-Bm
MetaBayes
MetanetBp

MetanetCm
FF-Sn
TasmSABp
TasmSASn
FF-BP
FF-Cm

A sequence of Nuclear Magnetic Resonance images of the breast region of different patients, which were not used for generating the teaching database are encoded as disclosed above with reference to Figure 1, and
 5 according to the pixel encoding method which is followed to encode the images used to create the teaching database for the processing algorithms. An example of these images is shown in Figure 8. The white ring denotes the presence of benign tumor tissue.

10 The identification vectors for the individual pixels are provided to the processing algorithm for the recognition of the tissue type reproduced thereby.

The algorithm assigns to the different identification vectors, hence to the corresponding
 15 pixels, the type of tissue represented thereby based on the teaching database.

The result thereof is displayed by appropriately and differentially coloring the pixels whereto the type of benign or malignant tumor tissue has been assigned.

20 Figure 9 shows a tissue type recognition result example referred to the image of Figure 8, in which the white outlined area had been recognized by visual analysis as representing the benign tumor tissue.

In Figure 9, the black screened white zone

represents the pixels whereto the processing algorithms assigned the type of benign tumor tissue.

The white encircled black zones denote the pixels whereto the processing algorithms assigned the type of malignant tumor tissue.

As compared with the recognition of malignant or benign tumor tissue types, obtained through visual analysis by a physician, for some pixels the algorithm provided non accordant indications, which have been temporarily classed as wrong.

Figure 5 shows both in a data table and in a chart the prediction reliability results of tissue type recognition obtained by processing with the different neural networks as listed above. The results obtained therefrom are expressed in terms of correct benign or malignant tumor tissue recognition percentage, of recognition sensitivity and of weighted and arithmetic correct recognition average, as well as absolute errors. The chart only shows the two tissue type recognition percentages and the errors.

The above description clearly shows the high tissue recognition reliability obtained by the processing algorithm, hence the high reliability provided by the processing method, when used as a diagnostic image analysis method, for recognizing and indicating the presence of types of tissues.

It is apparent that this method is not a pure diagnostic method, because, although the indications provided thereby are highly reliable, they can provide no diagnostic certainty nor substitute or prevent the

execution of additional specific analyses or examinations needed for a totally reliable diagnosis.

The method substantially provides support to diagnostic image reading and interpretation, aimed at
5 better location and recognition of specific tissue types represented in the images. The difficulties in reading and interpreting diagnostic images, whether of the MRI and ultrasound or radiographic imaging types are self-evident from Figure 8.

10 Obviously, a better result may be achieved by changing the encoding rule, i.e. by increasing the number of surrounding pixels around the pixel to be encoded, which are designed to form the components of the identification vector of said pixel to be encoded.

15 Better results were also achieved by increasing the number of the tissue types contained in the teaching database, as shown in the following examples.

EXAMPLE 2

20 The example 2 is similar to the example 1, an additional tissue type, i.e. normal tissue, being included in the recognition database. Hence, when the teaching database is generated, the encoded vectors for the pixels of the images are uniquely associated to one
25 of the tissue types represented thereby, i.e. benign tumor tissue, malignant tumor tissue or normal tissue.

This additional type allows to count on a greater number of pixels and corresponding identification vectors having a sure meaning. In the previous example,
30 these pixel identifying vectors and the corresponding

pixels have no meaning for the processing algorithm, whereas in this second example, the processing algorithm can assign an additional well-defined class or type of tissue.

5 This additional possibility provides an error reduction, since the algorithm may choose from three conditions or statuses to be assigned to the identification vectors.

10 The execution of the example is exactly as described and shown with reference to the previous example 1.

In this example, only the following neural networks were selected as processing algorithms:

PROCESSING ALGORITHM
MetaGen1
MetaGen
MetanetAf
MetaBayes
MetanetBp
MetanetCm
FF-Sn
FF-BP
FF-Cm

The numerical results are listed in Figure 6.

15 The recognition of tissue types by the different neural networks is more reliable.

EXAMPLE 3

The example 3 is similar to the above examples, but includes five tissue types, i.e.: benign tumor tissue, 20 malignant tumor tissue, normal tissue, muscular tissue

and image background.

The teaching database is generated as described above with reference to the previous examples and includes pixel identifying vectors, each being uniquely
5 assigned one of the above five types, i.e. the one represented by the respective pixel.

The tissue type recognition result per image pixel is shown in Figure 7 and for the following algorithms:

PROCESSING ALGORITHM
MetanetAf
MetaBayes
MetanetBp
MetanetCm
FF-Bm
FF-Sn
FF-Bp
FF-Cm
LDA

All these processing algorithms, except LDA, are
10 neural networks. LDA is a discriminating algorithm.

The result achieved thereby proves a very high tissue type recognition reliability. It shall be further noted that the discriminating algorithm also provides unexpected results in relation to the
15 capabilities thereof, when compared with normal conditions, although the results provided thereby are definitely lower than those obtained by neural networks.

The chart shows the errors for each different
20 algorithm.

Fig. 10 shows an example of result visualization by differentiated pixel coloring, depending on the different types recognized therefor, and with reference to the example of Figure 8.

5 The muscular tissue, the background, the normal tissue and the benign tumor tissue are properly recognized. The indications detected in Figure 9, regarding the example 1, in which in some small portions benign tumor tissue was wrongly located, have
10 disappeared herein.

Figures 11, 12 and 13 show an example of Nuclear Magnetic Resonance digital photographs of a breast region including a malignant tumor tissue, as highlighted by a white ring in Figure 11 and by a
15 corresponding partial enlarged view in Fig. 11.

Fig. 13 shows the result obtained in terms of recognition of the tissue types reproduced by the image pixels, thanks to the method of the invention, with the help of a neural network as a processing algorithm. The
20 teaching database is the same as in the example 3, including all five tissue types.

Correct recognition of malignant tumor, background, and muscular tissue is apparent from Fig. 13, which shows an example of recognition processing
25 result visualization.

As regards the described examples and the processing method of the invention, it shall be noted that, while identical frames are always used in the figures of the examples, the described method is not
30 necessarily limited to the type of frame. Thanks to the

fact that the encoding method according to the invention allows to account for the relation between the encoded pixel and the surrounding pixels, the teaching database actually allows to identify and
5 recognize the tissue type reproduced by a pixel of a digital image of the same anatomic region or possibly of different anatomic regions, regardless of the specific image frame.

This considerably simplifies imaging, as it does
10 not require images to be always acquired with the same frame.

It shall be further noted that the teaching database may dynamically grow by the addition of data gathered and confirmed through successive processing
15 procedures. In fact, once the tissue type assignment by the algorithm to a predetermined pixel and to the corresponding identification vector has been confirmed, the identification vector-tissue type pairs so formed may be themselves loaded in the teaching database which
20 grows with the use of the method, thereby making the processing algorithm increasingly expert and reducing the indecision or error margin.

It shall be additionally noted that, by simply adding a tissue type to be recognized and assigning it
25 to pixel identification vectors of images used to generate the teaching database, and which reproduce said tissue type, the recognition processing may be changed as regards the number of different tissue types to be recognized.

30 By totally changing the teaching database, it is

possible to recognize other tissue types in other types of subjects, and to correct or overlay images of the same subject, obtained with different techniques.

Image processing aimed at recognizing tissue types
5 or qualities is also possible by a pixel vector encoding method, which accounts for time variations of the pixel reproducing a specific object, i.e. the type according to the encoding example of Figure 3. As disclosed above, this encoding type allows to encode
10 pixels of image sequences.

In the diagnostic field, a tissue recognition method may be provided for moving subjects, such as in the case of ultrasound imaging or similar, of the heart.

15 Currently, upon the control of the electrocardiogram, sequences of ultrasound images of the heart region are acquired, to be further stored and displayed successively like in a movie sequence. The interpretation of the images within the sequence,
20 especially with reference to specific diseases is not easy and is relatively uncertain. Hence, thanks to the processing method as described with reference to the examples 1, 2 and 3, as combined with a method of encoding image pixels of image sequences of the same
25 type as described with reference to Figure 3, an automated analysis and recognition of the tissue types reproduced by the individual pixels of the image sequence may be performed, whereupon said types may be highlighted in a differentiated manner, over the
30 sequence images, to improve the viewing and locating

potential of the physician or the qualified personnel.

A similar application field for a combination of the processing method for tissue type recognition in digital or digitized image sequences with a pixel
5 encoding method for the pixels of said sequence images, respectively by an identification vector including, for each pixel, the values of the pixel to be encoded and of the pixels around it, of each image of the image sequence, consists in the recognition of tissues or
10 vascular or lymphatic flows in combination or not with the injection of contrast agents, as well as in the recognition and measurement of contrast agent perfusion.

In this case, the sequence of images acquired with
15 time after the injection of contrast agents is encoded with the method as described with reference to Figure 3. The teaching database for the processing algorithm includes behavior types, e.g. arterial blood flow or lymphatic or venous flow and stationary tissues and/or
20 tissues of vessel walls. Then, the recognition results are displayed e.g. by appropriately coloring the pixels relating to the different types.

In all the above cases, it is important to generate an adequate teaching database containing the
25 relevant object types or qualities for the specific examination.

It shall be noted that the same processing unit, i.e. the hardware wherein the processing software is loaded, may perform any of the above recognition
30 processing procedures, by simply providing the

processing software with the proper teaching database for the images to be processed and by obviously encoding the images to be processed.

Regarding image correction procedures, i.e. image
5 processing aimed at suppressing or recognizing artifacts and/or at recognizing defocused areas and at correcting them by focusing the image, the processing method of the invention does not substantially change.

As regards artifacts, a teaching database shall be
10 generated in which known images with or without artifacts are encoded, by assigning the artifact type to artifact-reproducing pixels and the correct pixel type to correct object reproducing pixels. Once an image or a sequence of images has been recognized, it
15 may be easily corrected by suppressing artifact-related pixels or by assigning to artifact-related pixels the tissue types or qualities which they might have with reference, for example, to surrounding pixels.

Defocusing may be corrected in a similar manner.

20 The processing method of the invention may be also advantageously used to generate images composed of individual images of the same subject as obtained by different techniques, e.g. Nuclear Magnetic Resonance imaging, ultrasound imaging and x-ray imaging.

25 In this case, the teaching database will contain pixel encoding vectors for all three images obtained with the three different techniques, with tissue types or qualities corresponding to said pixels being uniquely associated to said vectors. Hence, image
30 portions are uniquely associated to specific tissue

types and said well-defined portions may be displayed in overlaid positions or other combined arrangements within a single image.

5 An additional application of the inventive recognition method, in combination with imaging methods, particularly for diagnostic purposes, such as ultrasound or Nuclear Magnetic Resonance imaging methods, consists in that imaging is performed with less accurate but considerably faster imaging sequences
10 or techniques and that the displayed image is an image processed with the recognition method of this invention.

This allows to maintain high reliability levels in the recognition of the types or qualities reproduced by
15 the individual image pixels and to shorten the diagnostic imaging times. This arrangement may be very useful particularly for ultrasound or Nuclear Magnetic Resonance imaging, which require relatively long imaging times, in certain situations, and provides
20 apparent advantages.

The processing method of the invention also provides considerable advantages in the recognition of tissue types, like potentially diseased tissues, for example tumor tissues at very early stages. Currently,
25 x-ray mammography, for example, is performed with spatial resolutions of about 7 micron. Therefore, these images or the data associated thereto have such a resolution that different tissue types may be discriminated at very early growth levels, in groups of
30 a few cells. Nevertheless, the human eye only has a

spatial resolution of 100 micron. Hence the considerable imaging resolution cannot be currently used.

Conversely, the method of this invention does not
5 have any spatial resolution limits, except those possibly associated to image digitizing means.

Therefore, the spatial resolution limits of the human eye may be lowered by using appropriate image digitizing or digital sampling means, to potentially
10 reach the spatial resolution available at the imaging stage.

Therefore, by using the method of this invention, it is possible to generate a digitized virtual image which consists of a two-dimensional, three-dimensional
15 or multi-dimensional set, in which the virtual image is composed of image data for image unit dots relating to the spatial resolution below the one of the human eye.

The processing method essentially includes the same steps as described above, i.e. generating pixel
20 encoding vectors and quality and type recognition processing, particularly for tissues, as described above. As described herein, the different object types or object qualities may be highlighted by appropriately changing the aspect of the pixel related thereto, e.g.
25 by a suitable differentiated coloring arrangement.

From the set of pixel identification vectors, the pixel data matrix may be reconstructed, and said data may be used to control, for instance, a printer and/or a display screen.

30 The printer or display may be controlled in such a

manner as to allow the individual pixels to be also displayed at the resolution of the human eye, e.g. by using an image variation in which the data of each pixel of the high resolution image, i.e. having a resolution below that of the human eye is used to control a unit group of pixels of the display or printer, whose pixels take the same aspect as the corresponding pixel to be displayed. The image is inflated rather than enlarged, each high resolution pixel being represented by displaying a pixel sub-matrix which comprises a sufficient number of pixels to generate an image portion having a resolution of the same order of magnitude as the one of the human eye or higher.

15 In the case of an image composed of pixels whose size corresponds to a resolution of 7 micron, it is possible to form unit groups of $14 \times 14 = 196$ pixels, thus simulating a resolution of 98 micron.

Here, the 196 pixels of the unit group are controlled to assume the same aspect as assigned to the corresponding high definition pixel, thereby generating an image point which is visible to the human eye.

Obviously, the above displaying steps allow to generate unit groups of high definition pixels which may also have a greater or smaller number of pixels, substantially corresponding to a greater or smaller enlargement of the individual high resolution pixels.

The enlargement factor may be also user-preset by possibly allowing to delimit or define an image portion whereto the enlargement displaying step is to be

applied and to modify said image portion to be enlarged for successive and different enlargement steps with different enlargement and resolution factors.

5 The above description clearly shows that the recognition method according to this invention, based on the pixel encoding technique allows to process, evaluate or provide highly reliable indications, even regarding such information that is unrecognizable for the human eye.

10 Obviously, as for the method described above, multiple application fields may be provided, with particular reference to diagnostic image processing and to healthy or normal tissue or diseased tissue recognition, especially for benign and malignant tumor
15 tissues.

In the latter application field, the improvement as described above allows to analyze the tissue type and to obtain indications regarding the presence of benign or malignant tumor tissues at very early stages
20 which, at a resolution of 7 micron, are composed of a very small number of cells.